

LINKED-VIEW VISUALIZATION OF HIGH-DIMENSIONAL DATA IN GLUE

Alyssa A. Goodman

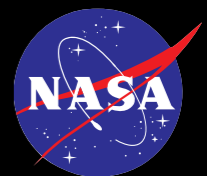
Harvard-Smithsonian Center for Astrophysics & Radcliffe Institute
with Chris Beaumont, Michelle Borkin, Penny Qian & Tom Robitaille



@aagie
@glueviz
@astrofrog



glueviz.org
github.com/glue-viz
Tom Robitaille, lead developer



James Webb
Space Telescope



“Linked Views”



Open Source Python, on GitHub

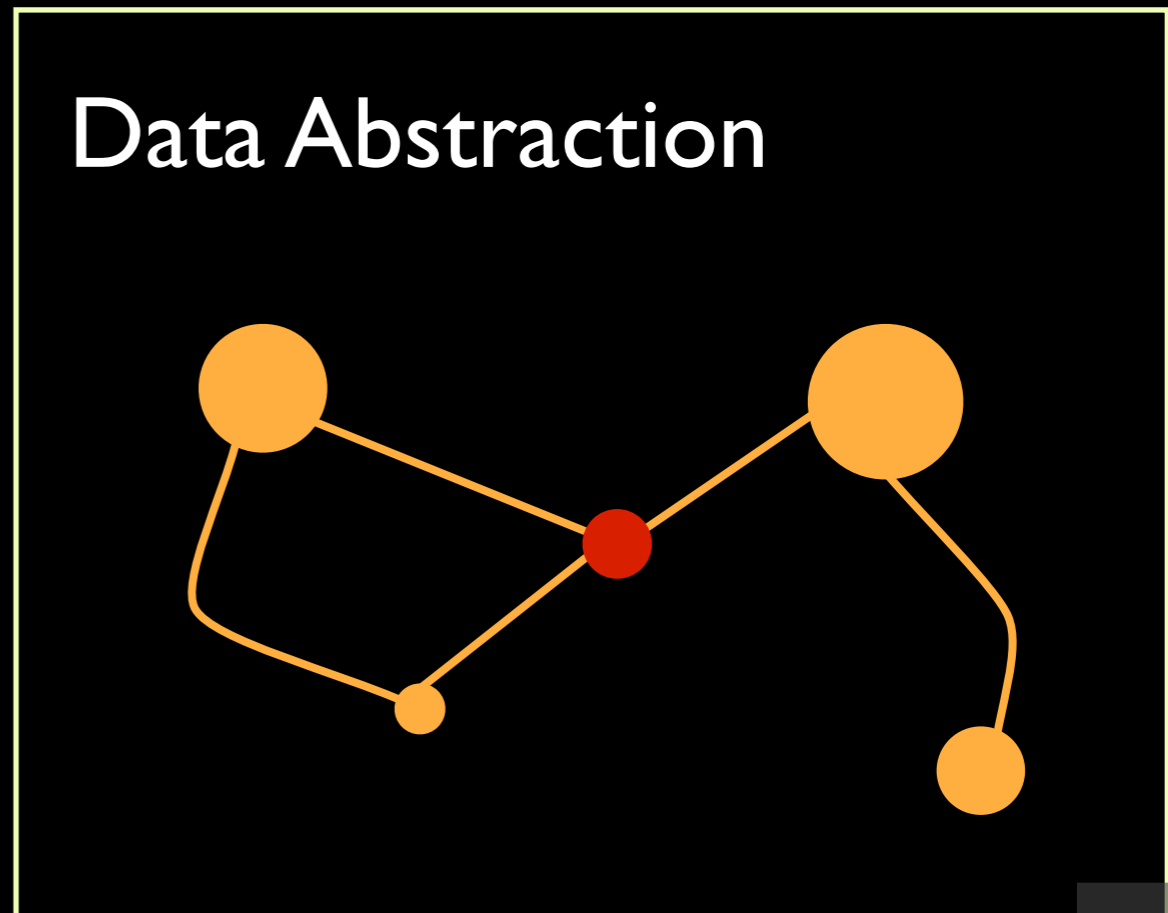
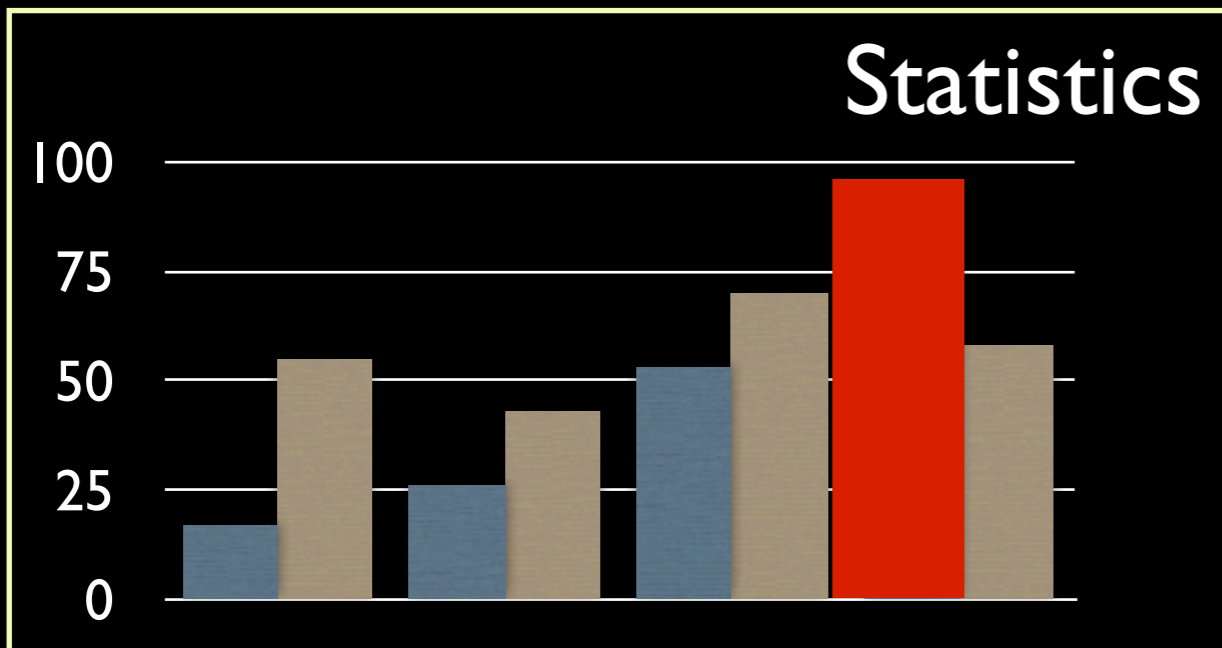
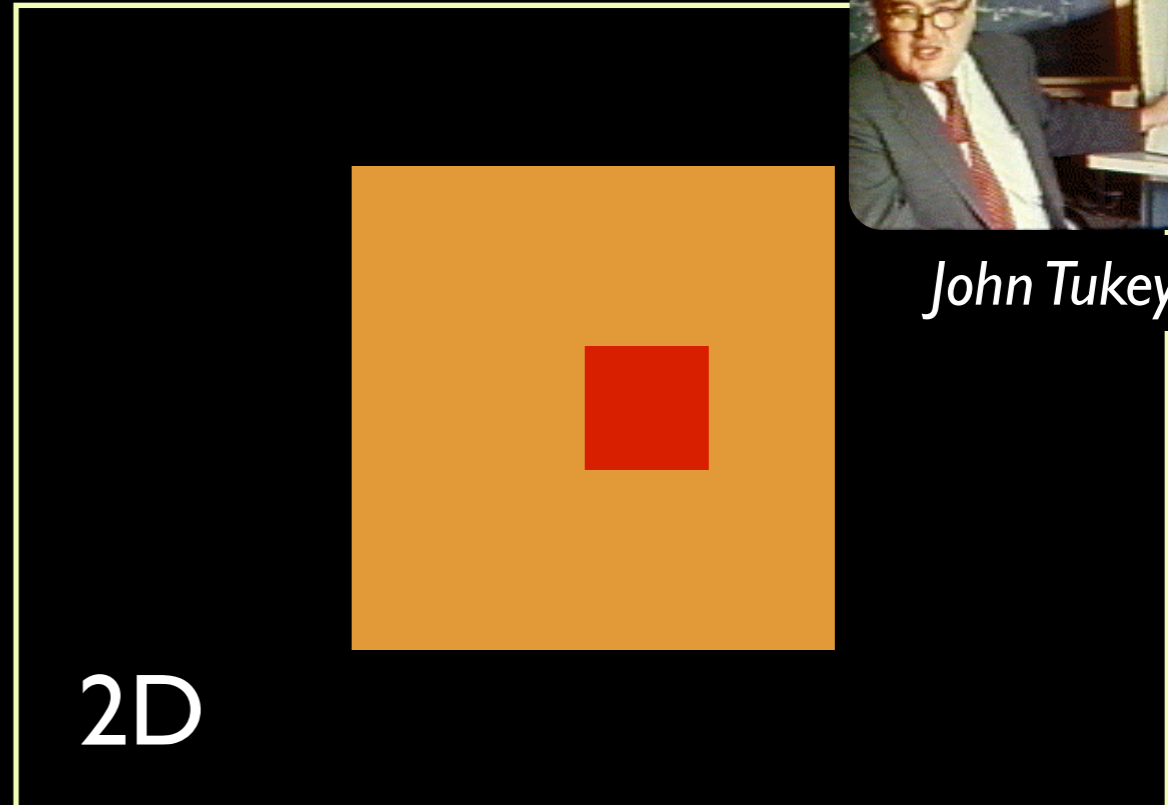
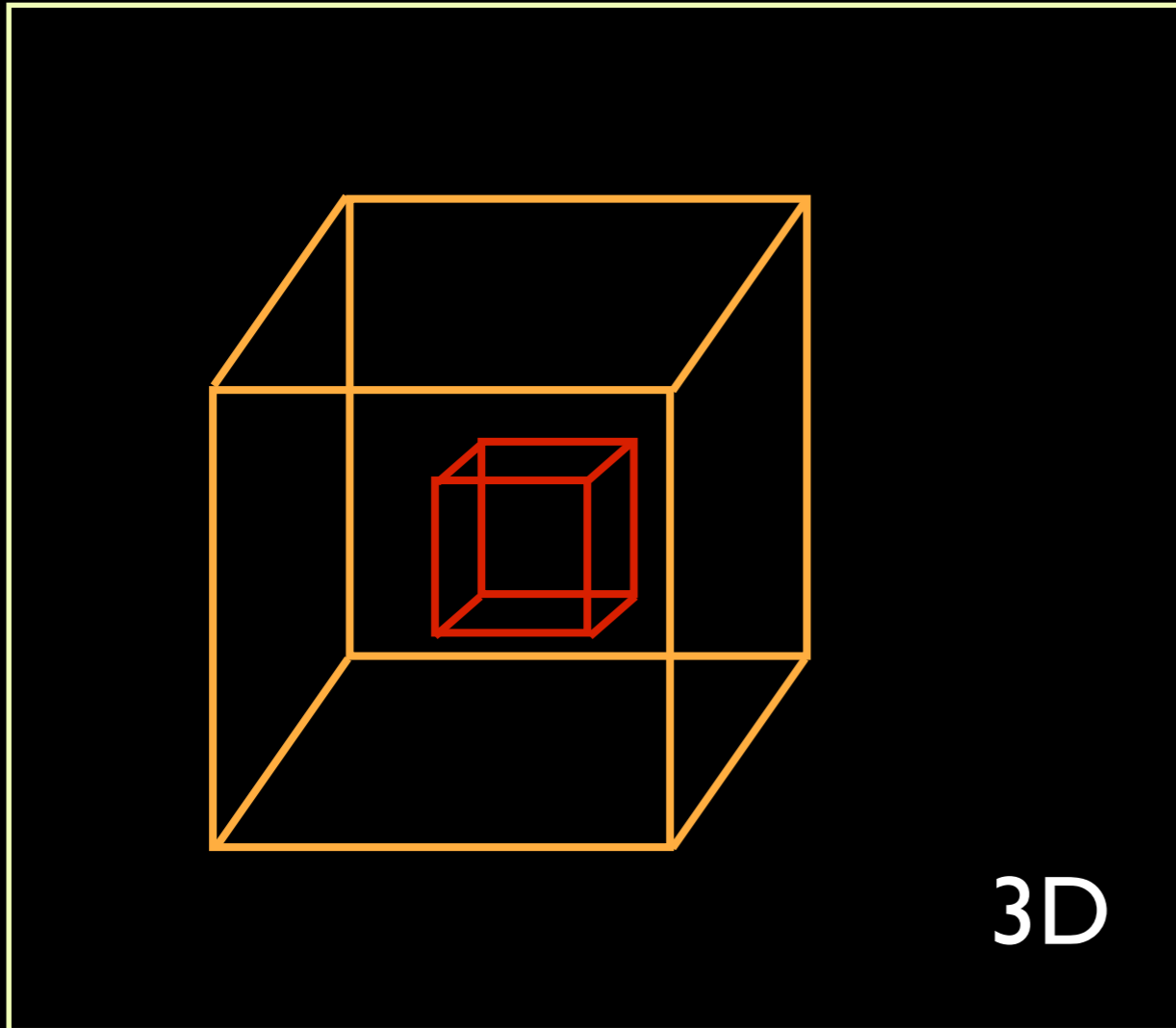


3D+

LINKED VIEWS OF HIGH-DIMENSIONAL DATA



John Tukey



figure, by M. Borkin, reproduced from Goodman 2012, "Principles of High-Dimensional Data Visualization in Astronomy"

JOHN TUKEY'S LEGACY



PRIM-9

PRIM-H

DataDesk®



XGobi

GGobi

RGGobi



Polaris



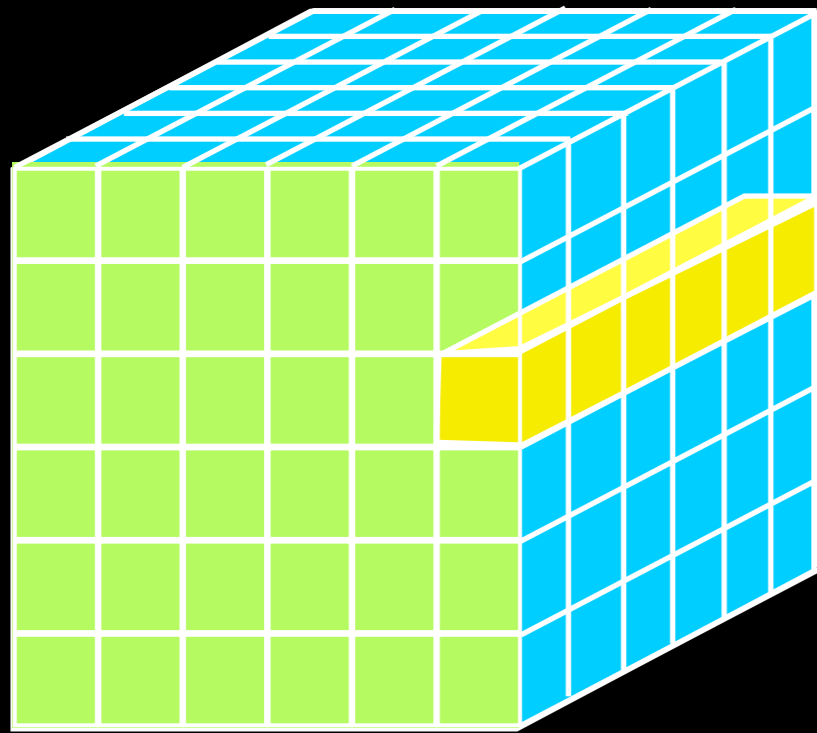
1970

1980

1990

2000

2010



DATA-DIMENSIONS-DISPLAY

1D: Columns = "Spectra", "SEDs" or "Time Series" (x-y Graphs)

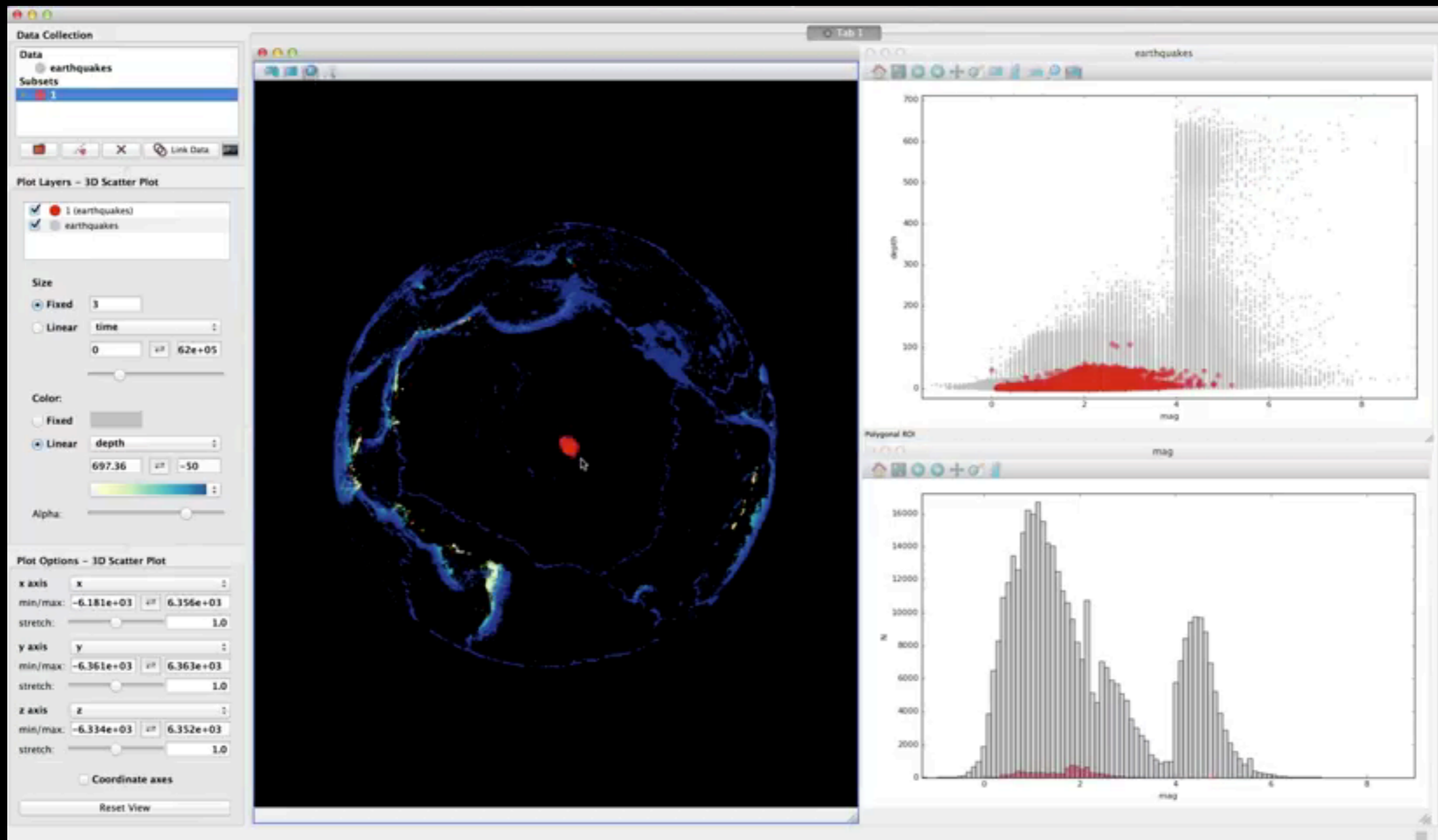
2D: Faces or Slices = "Images"

3D: Volumes = "3D Renderings", "2D Movies"

4D: Time Series of Volumes = "3D Movies"

LINKED VIEWS OF HIGH-DIMENSIONAL DATA (IN PYTHON)

GLUE



video by Tom Robitaille, lead glue developer
glue created by: C. Beaumont, M. Borkin, P. Qian, T. Robitaille, and A. Goodman, PI

"BUT WAIT, THERE'S MORE..."



dollars logo - Google Search

Building Custom Data Viewers — Glue 0.9.0 documentation

balzer82.g

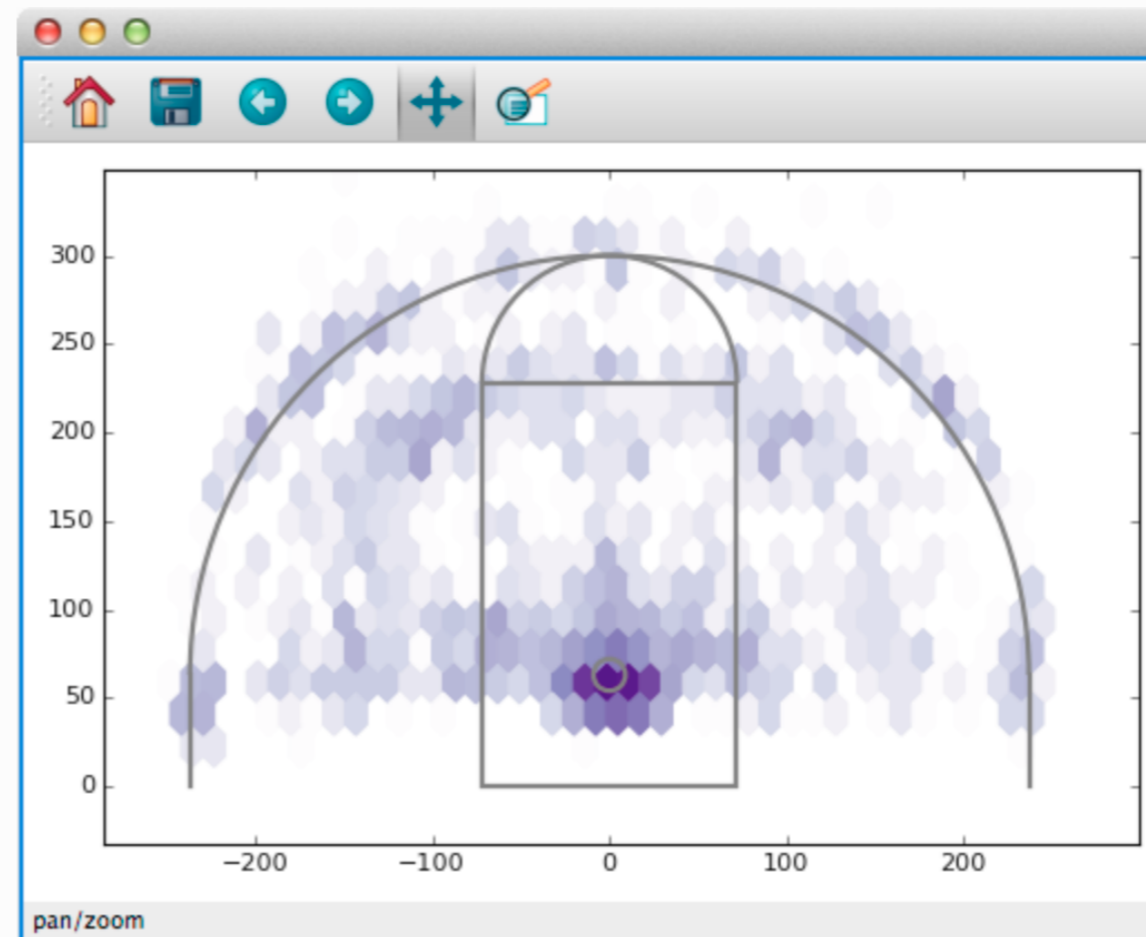
Glue

Search docs

[Docs](#) » Building Custom Data Viewers

[Edit on GitHub](#)

Building Custom Data Viewers



Glue's standard data viewers (scatter plots, images, histograms) are useful in a wide variety of data exploration settings. However, they represent a *tiny* fraction of the ways to view a particular dataset. For this reason, Glue provides a simple mechanism for creating custom visualizations using `matplotlib`.

Creating a `custom data viewer` requires writing a little bit of `Matplotlib` code but involves little to no GUI programming. The next several sections illustrate how to build a custom data viewer by

- Installing Glue
- Getting started
- User Interface Guide
- 3D viewers in Glue
 - Using the IPython terminal in Glue
- Working with Data objects
- Starting Glue from Python
- Configuring Glue via a startup file
- Customizing your Glue environment
- Programmatically configuring plots
- Building Custom Data Viewers**
 - The Goal: Basketball Shot Charts
 - Shot Chart Version 1: Heatmap and plot
 - Shot Chart Version 2: Court markings
 - Shot Chart Version 3: Widgets
 - Shot Chart Version 4: Selection
 - Viewer Subclasses
 - Valid Function Arguments
 - UI Elements
 - Other Guidelines

Watching data for changes

[Read the Docs](#)

v: stable

"BUT WAIT, THERE'S MORE..."



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balzer82.g

Glue

Search docs

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- Watching data for changes

Read the Docs

v: stable

Docs » Building Custom Data Viewers

"cuts" along arbitrary paths

Building Custom Data Viewers

flood-fill selection (2D, 3D)

export to d3po, plotly

custom viewers (e.g. GIS, WorldWide Telescope, Super Mario)

plot manipulation/customization (via Matplotlib)

flexible import/export

saved sessions (.glu)

Anaconda Navigator install/upgrade



Yes, please do go start adding code now, at github.com/glue-viz.

Glue's standard data viewers (scatter plots, images, histograms) are useful in a wide variety of data exploration settings. However, they represent a *tiny* fraction of the ways to view a particular dataset.

For this reason, Glue provides a simple mechanism for creating custom visualizations using

Creating a custom data viewer requires writing a little bit of Matplotlib code but involves little to no GUI programming. The next several sections illustrate how to build a custom data viewer by

INTEGRATION

The screenshot shows a dashboard with a sidebar on the left containing 'Home', 'Environments', 'Learning', and 'Community'. The main area is titled 'My Applications' and contains seven application cards:

- glueviz** (0.8.2): Multidimensional data visualization across files. Explore relationships within and among related datasets. [Launch]
- jupyter notebook** (4.2.1): Web-based, interactive computing notebook environment. Edit and run human-readable docs while describing the data analysis. [Launch]
- qtconsole** (4.2.1): PyQt GUI that supports inline figures, proper multiline editing with syntax highlighting, graphical calltips, and more. [Launch]
- spyder** (2.3.8): Scientific PYTHON Development EnviRonment. Powerful Python IDE with advanced editing, interactive testing, debugging and introspection features. [Launch]
- vaex** (1.0.0b3): [Launch]
- orange3** (3.3.8): [Install]
- xonsh** (0.4.7): [Install]

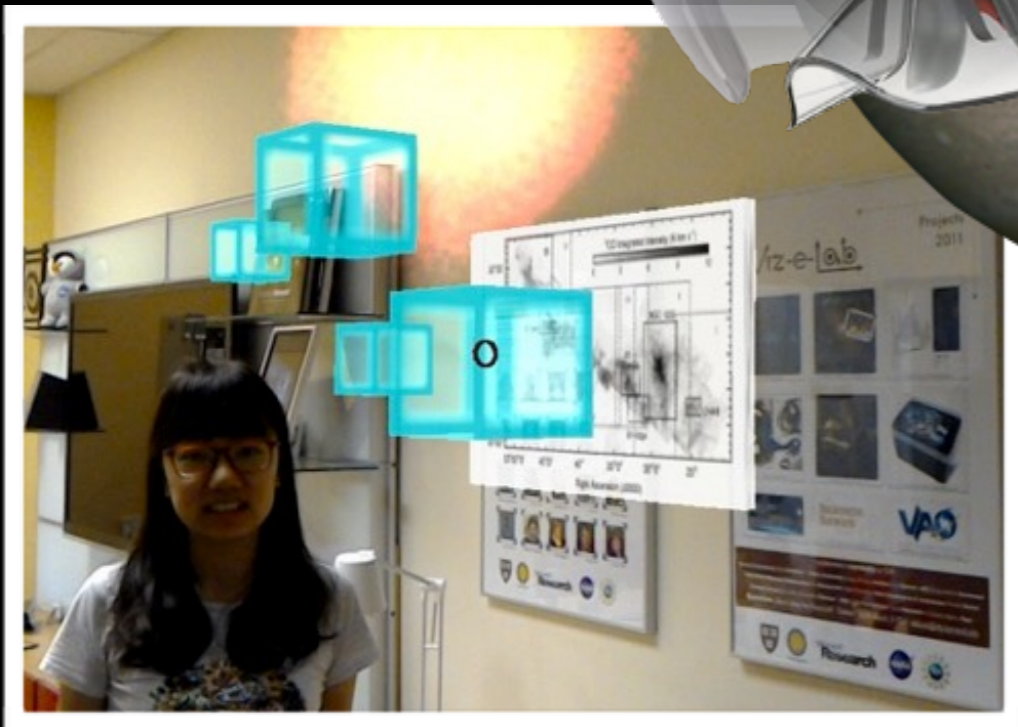
At the bottom of the sidebar are links for 'Documentation', 'Developer Blog', and 'Feedback', along with social media icons for Twitter, YouTube, and GitHub.

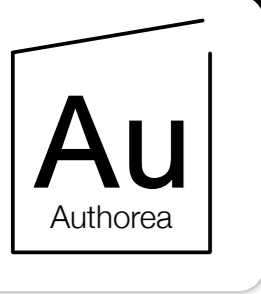
A vertical stack of logos for various technologies:

- python™
- SciPy
- jupyter
- astroPy: A Community Python Library for Astronomy
- JavaScript
- d3.js
- A grey envelope icon with a question mark inside.
- HTML5



THE CHALLENGE OF 3D SELECTION





The "Paper" of the Future

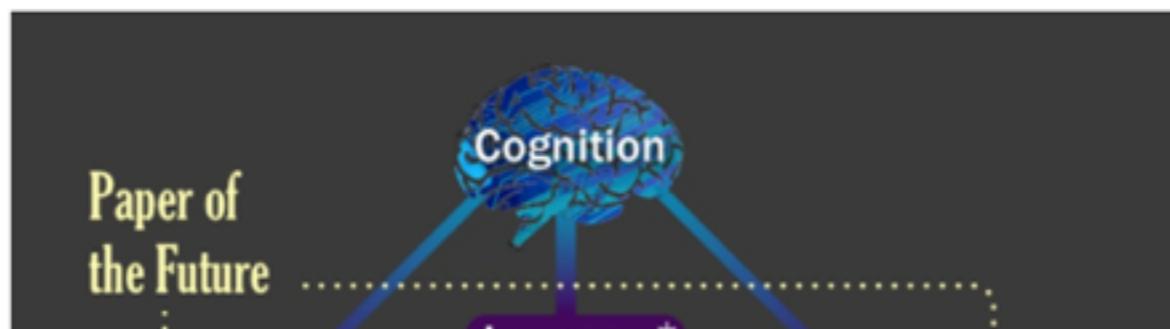
Alyssa Goodman, Josh Peek, Alberto Accomazzi, Chris Beaumont, Christine L. Borgman, How-Huan Hope Chen, Merce Crosas, Christopher Erdmann, August Muench, Alberto Pepe, Curtis Wong

A 5-minute video demonstration of this paper is available at [this YouTube link](#).

1 Preamble

A variety of research on human cognition demonstrates that humans learn and communicate best when more than one processing system (e.g. visual, auditory, touch) is used. And, related research also shows that, no matter how technical the material, most humans also retain and process information best when they can put a narrative "story" to it. So, when considering the future of scholarly communication, we should be careful not to do blithely away with the linear narrative format that articles and books have followed for centuries: instead, we should enrich it.

Much more than text is used to communicate in Science. Figures, which include images, diagrams, graphs, charts, and more, have enriched scholarly articles since the time of Galileo, and ever-growing volumes of data underpin most scientific papers. When scientists communicate face-to-face, as in talks or small discussions, these figures are often the focus of the conversation. In the best discussions, scientists have the ability to manipulate the figures, and to access underlying data, in real-time, so as to test out various what-if scenarios, and to explain findings more clearly. **This short article explains—and shows with demonstrations—how scholarly "papers" can morph into long-lasting rich records of scientific discourse, enriched with deep data and code linkages, interactive figures, audio, video, and commenting.**



3

Konrad Hinsen 3 days ago · Public

Many good suggestions, but if the goal is "long-lasting rich records of scientific discourse", a more careful and critical attitude towards electronic artifacts is appropriate. I do see it concerning videos, but not a word on the much more critical situation in software. Archiving source code is not sufficient: all the dependencies, plus the complete build environment, would have to be conserved as well to make things work a few years from now. An "executable figure" in the form of an IPython notebook will...

[more](#)

2

Merce Crosas 3 days ago · Public

Konrad, good points; this has been a concern for the community working on reproducibility. Regarding data repositories, Dataverse handles long-term preservation and access of data files in the following way: 1) for some data files that the repository recognizes (such as R Data, SPSS, STATA), which depend on a statistical package, the system converts them into a preservation format (such as a tab/CSV format). Even though the original format is also saved and can be accessed, the new preservation format gua...

[more](#)

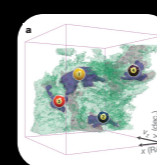
0

Konrad Hinsen 1 day ago · Public

That sounds good. I hope more repositories will follow the example of Dataverse. Figshare in particular has a very different attitude, encouraging researchers to deposit as much as possible. That's perhaps a good strategy to change habits, but in the long run it could well backfire when people find out in a few years that 90% of those deposits have become useless.

Christine L. Borgman 4 months ago · Private

"publications"



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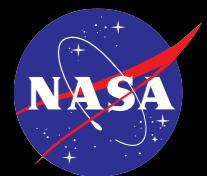
Harvard-Smithsonian Center for Astrophysics & Radcliffe Institute
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@aagie
@glueviz
@astrofrog



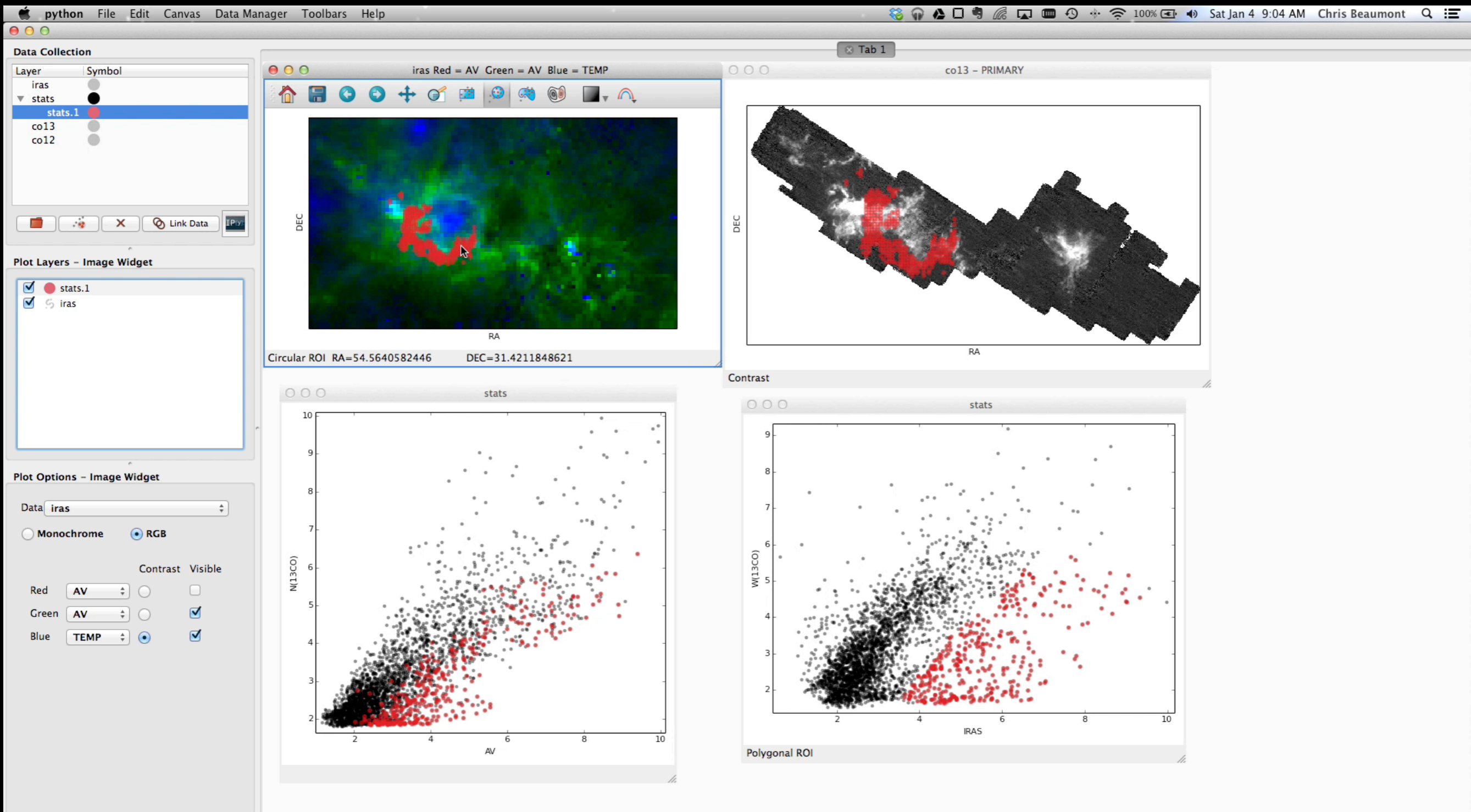
glueviz.org
github.com/glue-viz
Tom Robitaille, lead developer



James Webb
Space Telescope

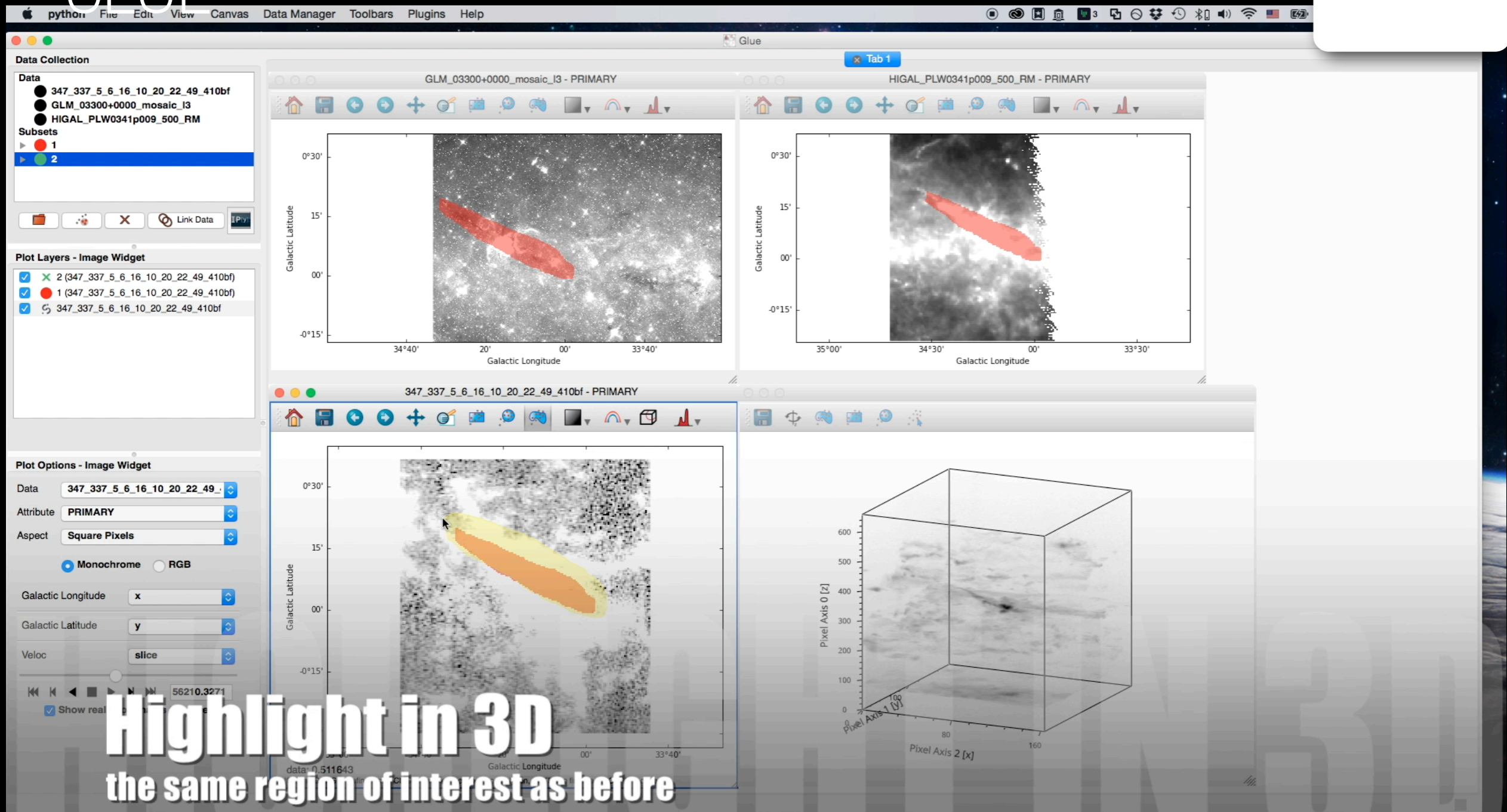
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GLUE



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GLUE

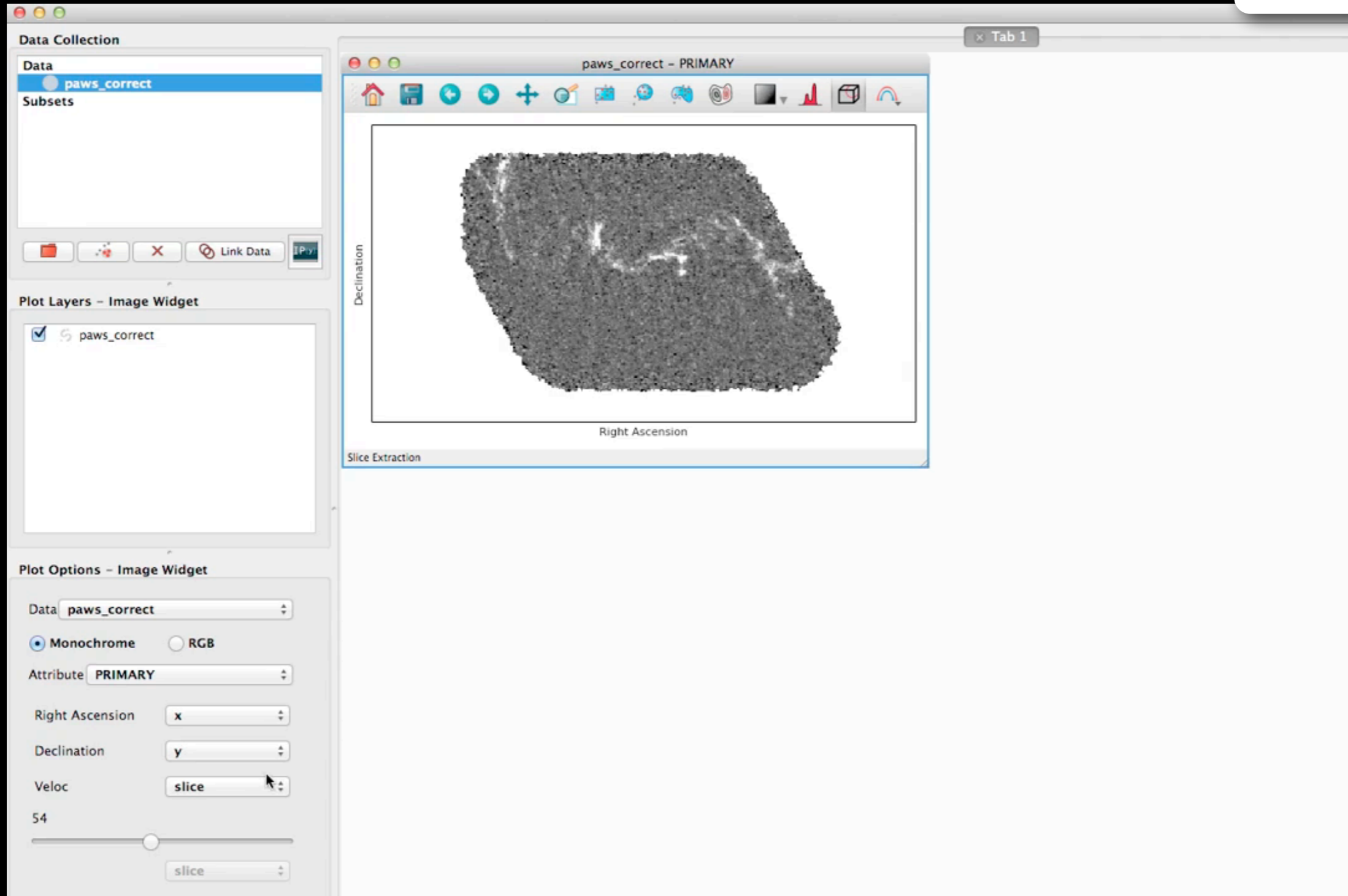


Highlight in 3D
the same region of interest as before

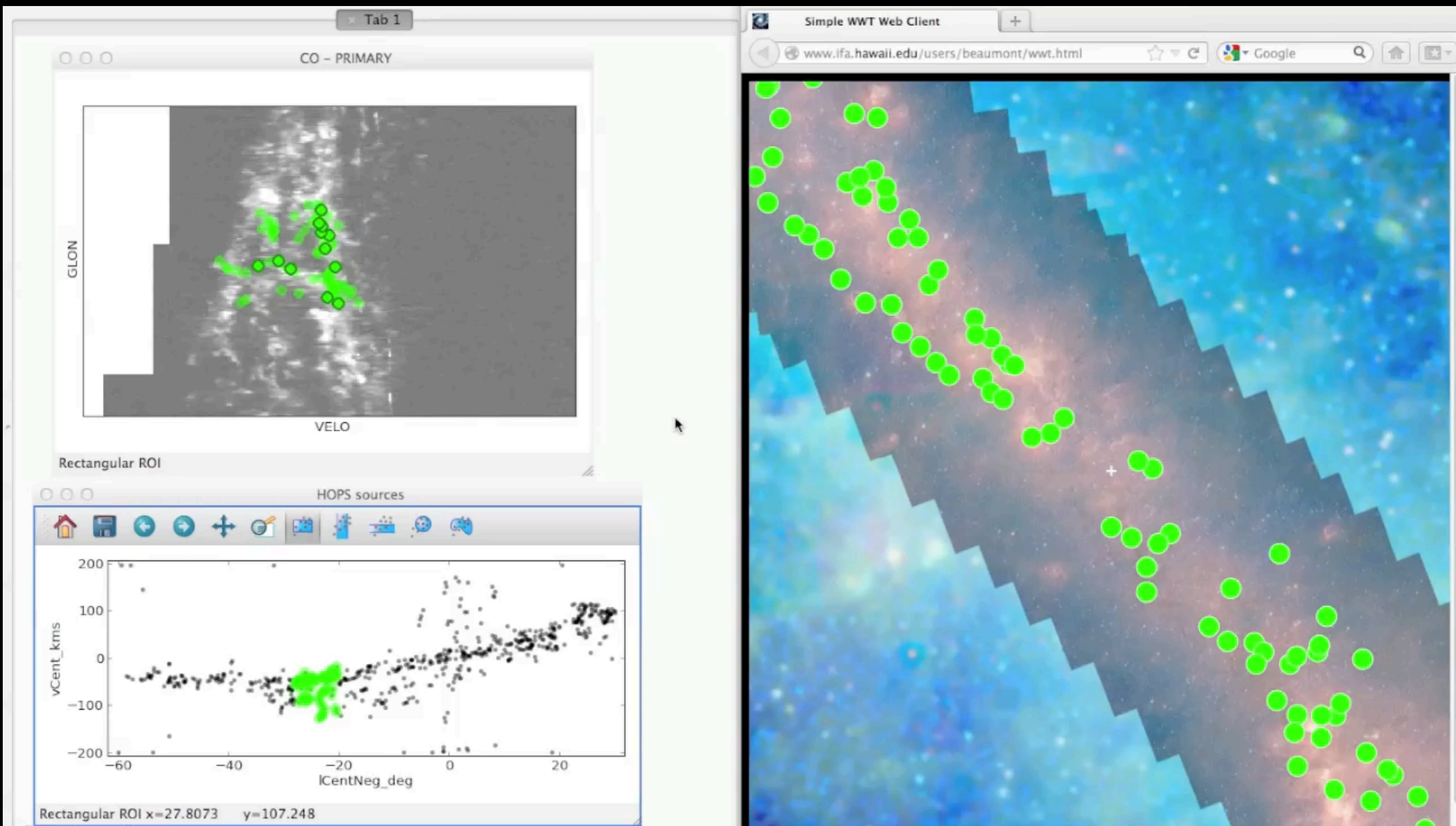
video by Penny Qian, with Catherine Zucker, graduate students
glue created by: C. Beaumont, M. Borkin, P. Qian, T. Robitaille, and A. Goodman, PI

LINKED VIEWS OF HIGH-DIMENSIONAL DATA (IN PYTHON)

GLUE



video by Chris Beaumont, glue developer
glue created by: C. Beaumont, M. Borkin, P. Qian, T. Robitaille, and A. Goodman, PI



Video courtesy of Chris Beaumont

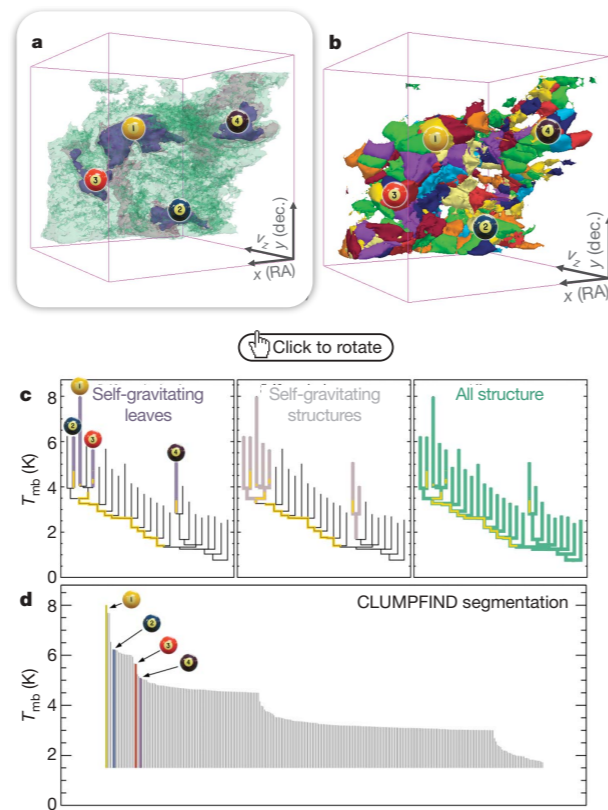


Figure 2 | Comparison of the 'dendrogram' and 'CLUMPFIND' feature-identification algorithms as applied to ^{13}CO emission from the L1448 region of Perseus. **a**, 3D visualization of the surfaces indicated by colours in the dendrogram shown in **c**. Purple illustrates the smallest scale self-gravitating structures in the region corresponding to the leaves of the dendrogram; pink shows the smallest surfaces that contain distinct self-gravitating leaves within them; and green corresponds to the surface in the data cube containing all the significant emission. Dendrogram branches corresponding to self-gravitating objects have been highlighted in yellow over the range of T_{mb} (main-beam temperature) test-level values for which the virial parameter is less than 2. The x - y locations of the four 'self-gravitating' leaves labelled with billiard balls are the same as those shown in Fig. 1. The 3D visualizations show position-position-velocity (p - p - v) space. RA, right ascension; dec., declination. For comparison with the ability of dendrograms (**c**) to track hierarchical structure, **d** shows a pseudo-dendrogram of the CLUMPFIND segmentation (**b**), with the same four labels used in Fig. 1 and in **a**. As 'clumps' are not allowed to belong to larger structures, each pseudo-branch in **d** is simply a series of lines connecting the maximum emission value in each clump to the threshold value. A very large number of clumps appears in **b** because of the sensitivity of CLUMPFIND to noise and small-scale structure in the data. In the online PDF version, the 3D cubes (**a** and **b**) can be rotated to any orientation, and surfaces can be turned on and off (interaction requires Adobe Acrobat version 7.0.8 or higher). In the printed version, the front face of each 3D cube (the 'home' view in the interactive online version) corresponds exactly to the patch of sky shown in Fig. 1, and velocity with respect to the Local Standard of Rest increases from front (-0.5 km s^{-1}) to back (8 km s^{-1}).

data, CLUMPFIND typically finds features on a limited range of scales, above but close to the physical resolution of the data, and its results can be overly dependent on input parameters. By tuning CLUMPFIND's two free parameters, the same molecular-line data set⁸ can be used to show either that the frequency distribution of clump mass is the same as the initial mass function of stars or that it follows the much shallower mass function associated with large-scale molecular clouds (Supplementary Fig. 1).

Four years before the advent of CLUMPFIND, 'structure trees'⁹ were proposed as a way to characterize clouds' hierarchical structure

using 2D maps of column density. With the help of 2D work as inspiration, we have developed a structure-identification algorithm that abstracts the hierarchical structure of a data cube into a form that is an easily visualized representation called a dendrogram. This algorithm, well developed in other data-intensive applications, has been applied to the application of tree methodologies so far, and almost exclusively within the area of astronomy. 'merger trees' are being used with increasing frequency.

Figure 3 and its legend explain the dendrogram process schematically. The dendrogram of a data cube is determined almost entirely by the choice of a test level, and its sensitivity to algorithm parameters. The dendrogram is possible on paper and 2D screen, and its construction is data (see Fig. 3 and its legend) and its construction is a process, which eliminates dimensions, preserving all information. The dendrogram is a hierarchical tree diagram. Numbered 'billiard ball' labels are used to track features between a 2D map and a sorted dendrogram.

A dendrogram of a spectrum is a hierarchical tree diagram of key physical properties. The dendrogram is a hierarchical tree diagram of surfaces, such as radius (R), velocity dispersion (σ_v), and luminosity (L). The volumes can have any shape, and the dendrogram is a hierarchical tree diagram of the significance of the especially elongated features.

(Fig. 2a). The luminosity is an approximate proxy for mass, such that $M_{\text{lum}} = X_{13\text{CO}} L_{13\text{CO}}$, where $X_{13\text{CO}} = 8.0 \times 10^{20} \text{ cm}^{-2} \text{ K}^{-1} \text{ s}$ (ref. 15; see Supplementary Methods and Supplementary Fig. 2). The derived values for size, mass and velocity dispersion can then be used to estimate the role of self-gravity at each point in the hierarchy, via calculation of an 'observed' virial parameter, $\alpha_{\text{obs}} = 5\sigma_v^2 R / GM_{\text{lum}}$. In principle, extended portions of the tree (Fig. 2, yellow highlighting) where $\alpha_{\text{obs}} < 2$ (where gravitational energy is comparable to or larger than kinetic energy) correspond to regions of p - p - v space where self-gravity is significant. As α_{obs} only represents the ratio of kinetic energy to gravitational energy at one point in time, and does not explicitly capture external over-pressure and/or magnetic fields¹⁶, its measured value should only be used as a guide to the longevity (boundedness) of any particular feature.

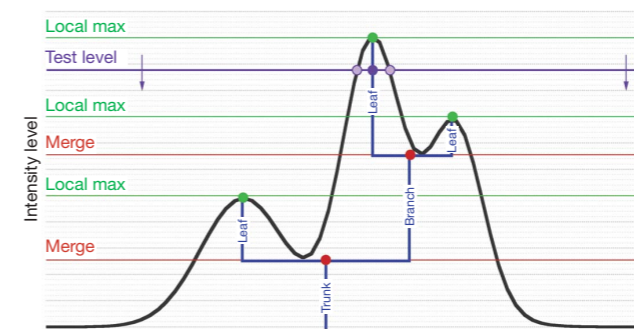
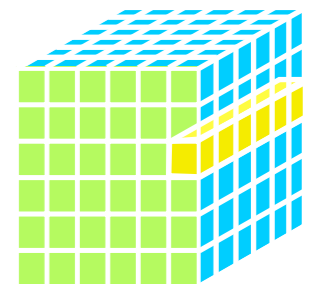
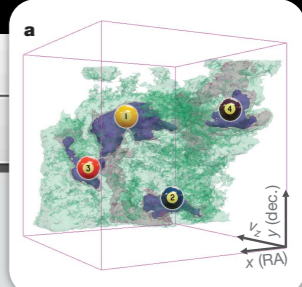


Figure 3 | Schematic illustration of the dendrogram process. Shown is the construction of a dendrogram from a hypothetical one-dimensional emission profile (black). The dendrogram (blue) can be constructed by 'dropping' a test constant emission level (purple) from above in tiny steps (exaggerated in size here, light lines) until all the local maxima and mergers are found, and connected as shown. The intersection of a test level with the emission is a set of points (for example the light purple dots) in one dimension, a planar curve in two dimensions, and an isosurface in three dimensions. The dendrogram of 3D data shown in Fig. 2c is the direct analogue of the tree shown here, only constructed from 'isosurface' rather than 'point' intersections. It has been sorted and flattened for representation on a flat page, as fully representing dendrograms for 3D data cubes would require four dimensions.

Goodman et al. 2009, Nature,
cf. Fluke et al. 2009

2009
3D PDF
HIGH-DIMENSIONAL
DATA IN A
"PAPER"





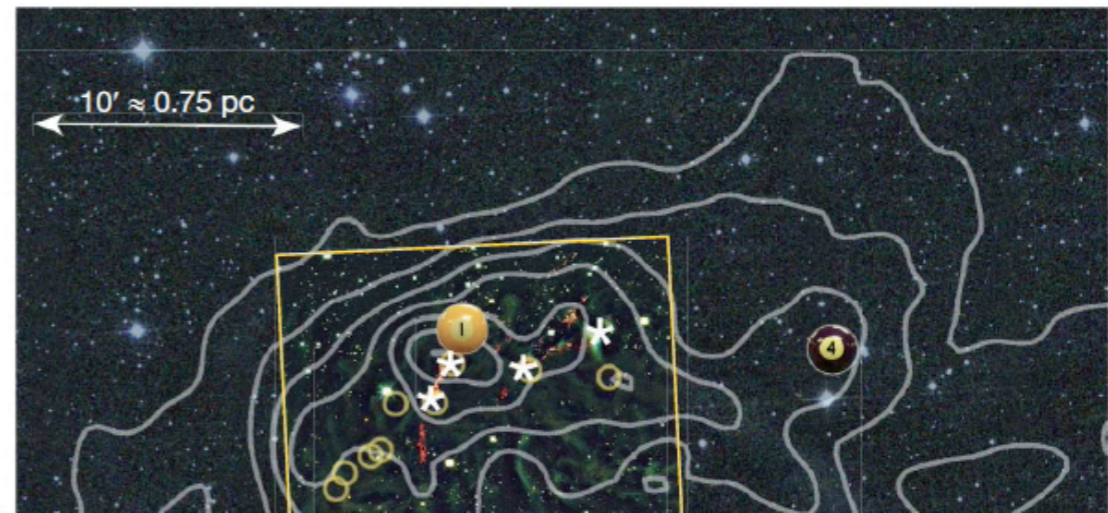
LETTERS

A role for self-gravity at multiple length scales in the process of star formation

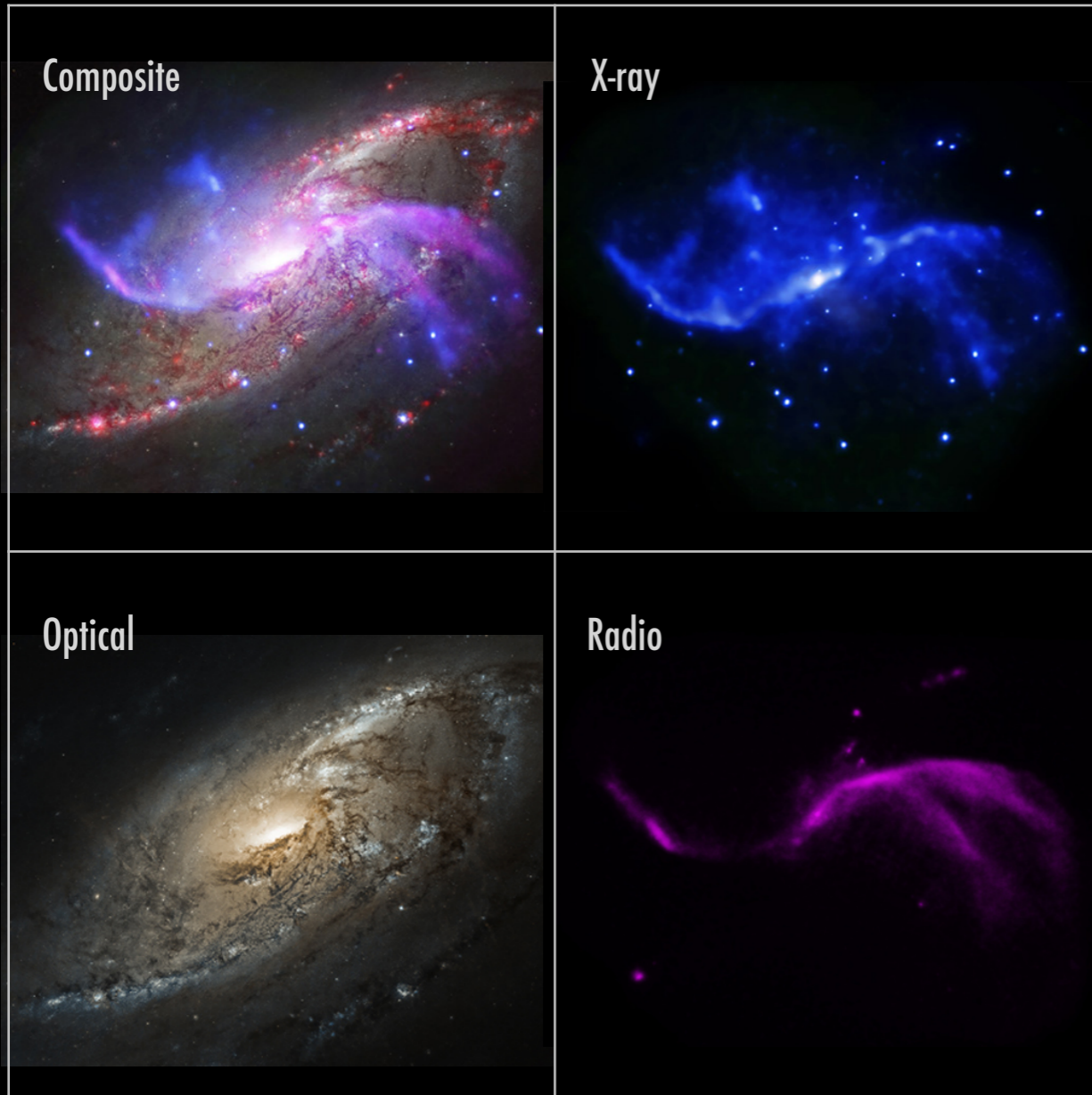
Alyssa A. Goodman^{1,2}, Erik W. Rosolowsky^{2,3}, Michelle A. Borkin^{1†}, Jonathan B. Foster², Michael Halle^{1,4}, Jens Kauffmann^{1,2} & Jaime E. Pineda²

Self-gravity plays a decisive role in the final stages of star formation, where dense cores (size ~ 0.1 parsecs) inside molecular clouds collapse to form star-plus-disk systems¹. But self-gravity's role at earlier times (and on larger length scales, such as ~ 1 parsec) is unclear; some molecular cloud simulations that do not include self-gravity suggest that 'turbulent fragmentation' alone is sufficient to create a mass distribution of dense cores that resembles, and sets, the stellar initial mass function². Here we report a 'dendrogram' (hierarchical tree-diagram) analysis that reveals that self-gravity plays a significant role over the full range of possible scales traced by ¹³CO observations in the L1448 molecular cloud, but not everywhere in the observed region. In particular, more than 90 per cent of the compact 'pre-stellar cores' traced by peaks of dust emission³ are projected on the sky within one of the dendrogram's self-gravitating 'leaves'. As these peaks mark the locations of already-forming stars, or of those probably about to form, a self-gravitating cocoon seems a critical condition for their exist-

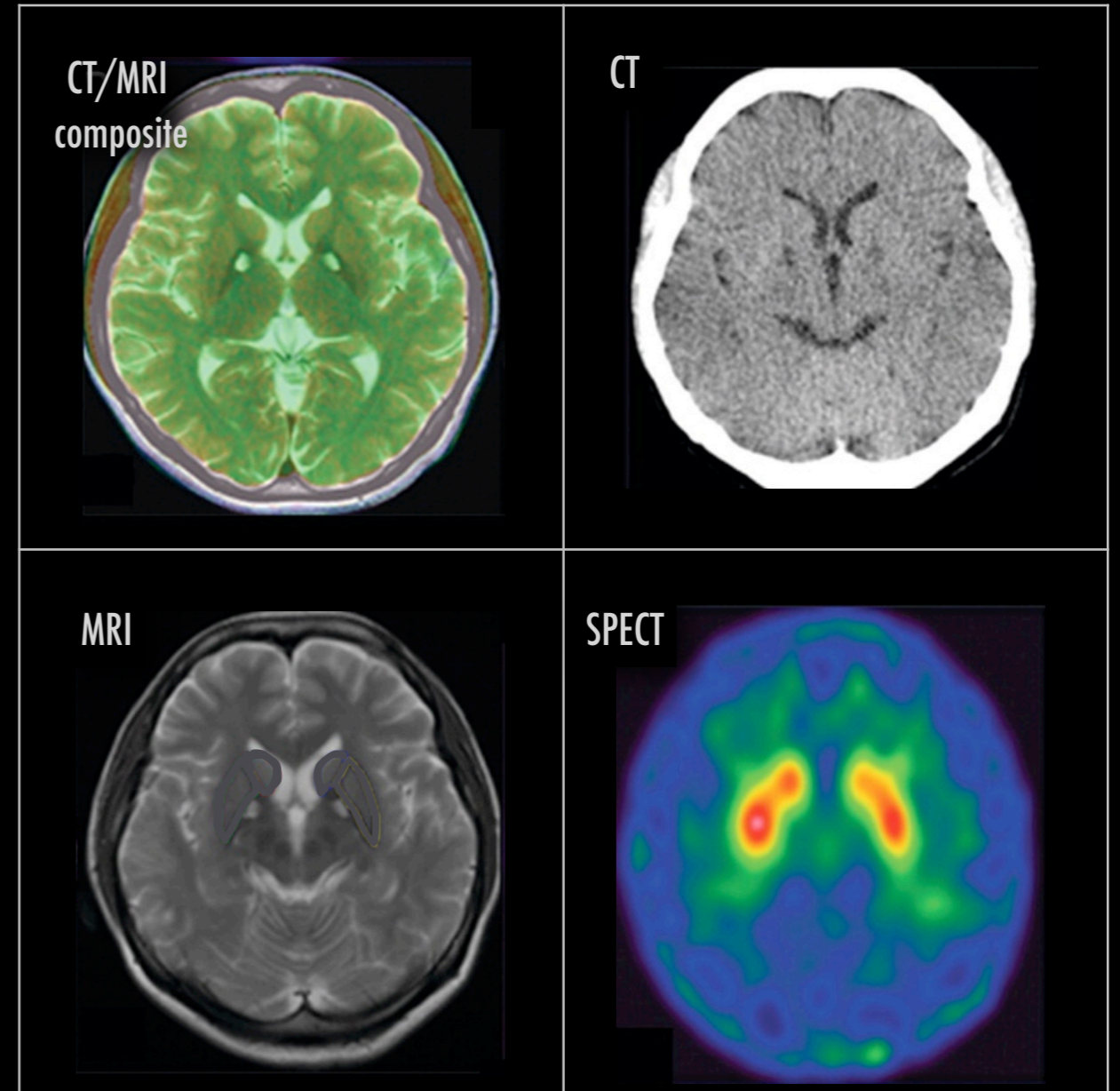
overlapping features as an option, significant emission found between prominent clumps is typically either appended to the nearest clump or turned into a small, usually 'pathological', feature needed to encompass all the emission being modelled. When applied to molecular-line



ASTRONOMICAL MEDICINE



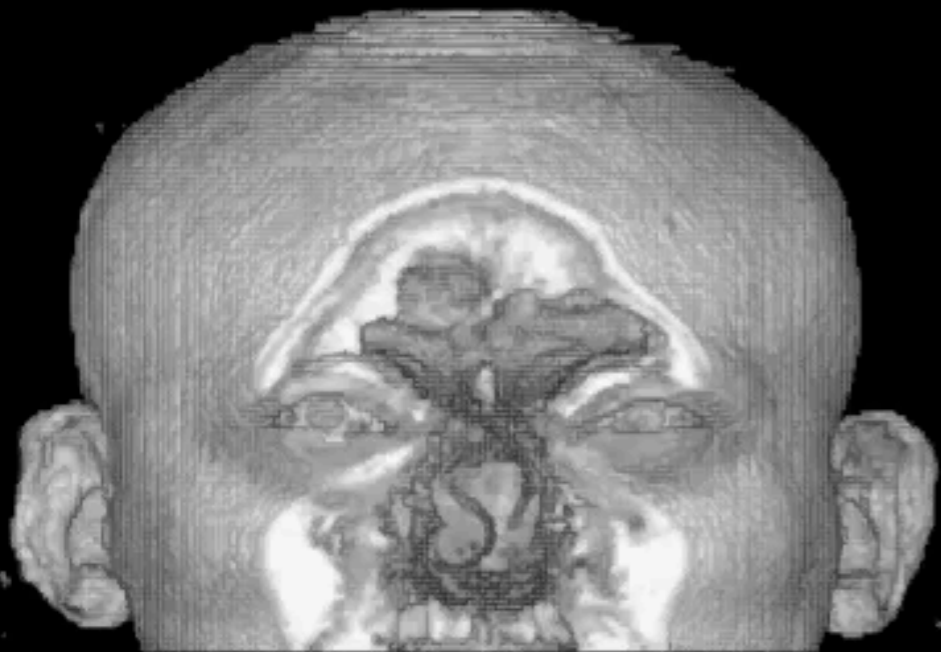
chandra.harvard.edu/photo/2014/m106/



Chang, et al. 2011, brain.oxfordjournals.org/content/134/12/3632

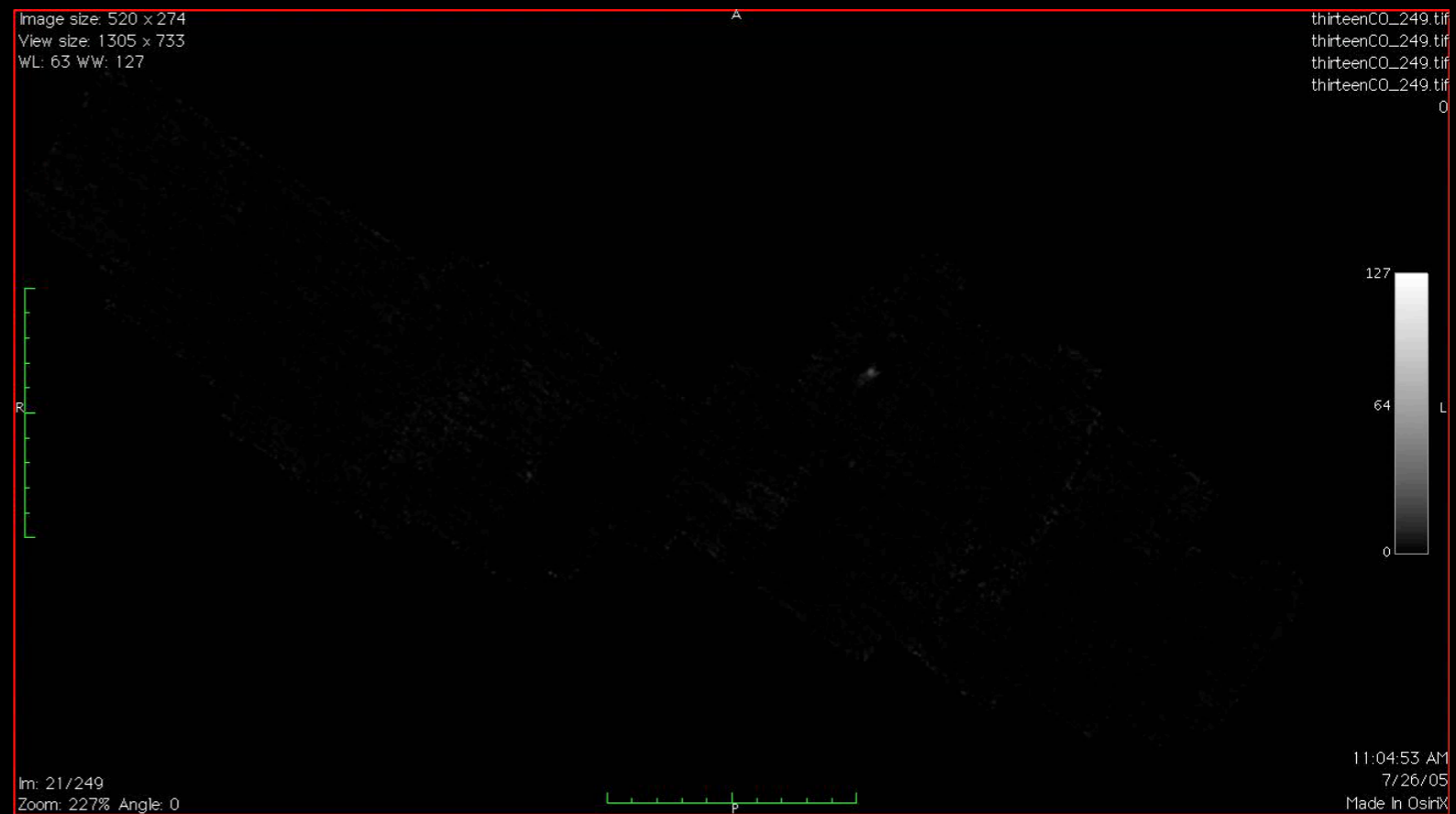
ASTRONOMICAL MEDICINE

"KEITH"



"z" is depth into head






"PERSEUS"

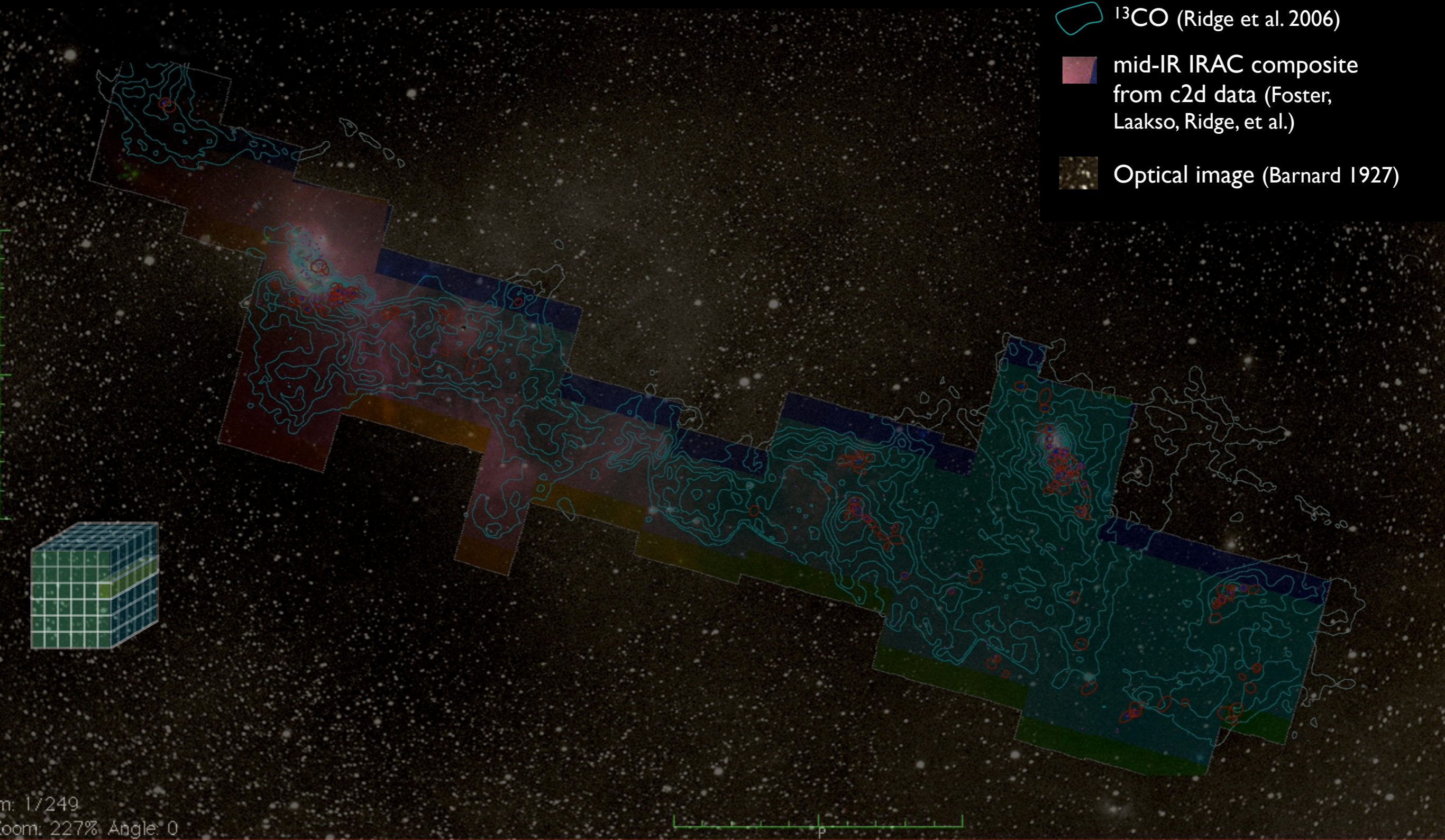


"z" is line-of-sight velocity

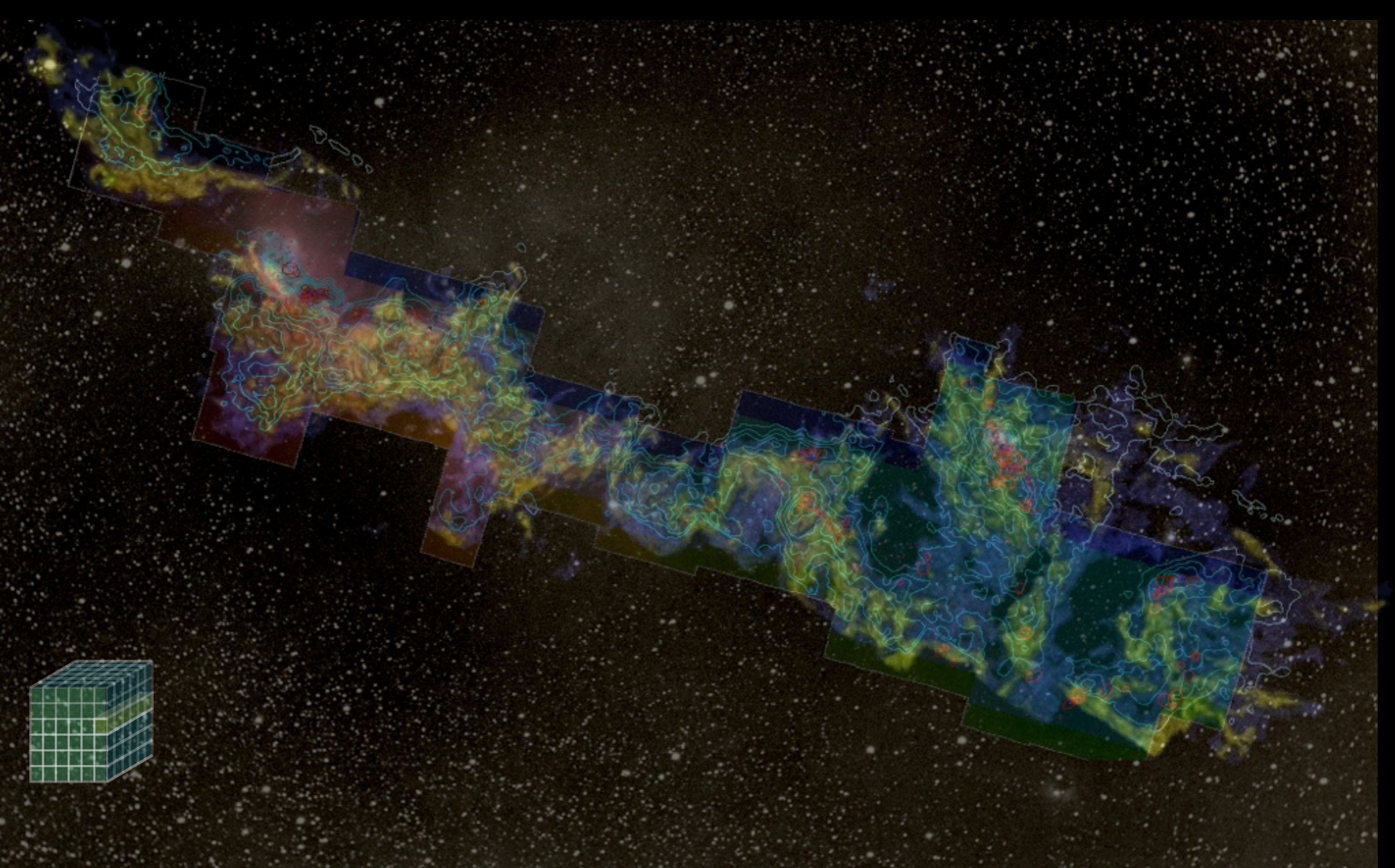
Image size: 520 x 274
View size: 1305 x 733
W/L: 63 WW: 127

ASTRONOMICAL MEDICINE

-  mm peak (Enoch et al. 2006)
-  sub-mm peak (Hatchell et al. 2005, Kirk et al. 2006)
-  ^{13}CO (Ridge et al. 2006)
-  mid-IR IRAC composite from c2d data (Foster, Laakso, Ridge, et al.)
-  Optical image (Barnard 1927)



m: 1/249
Zoom: 227% Angle: 0



3D Viz made with VolView